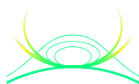


An efficient adjoint computation for flow control problems

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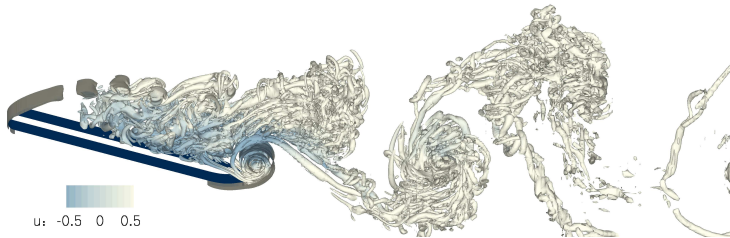
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Outline

- 1 Motivation
- 2 Semtex
- 3 Adjoint calculation
- 4 Checkpointing
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Motivation



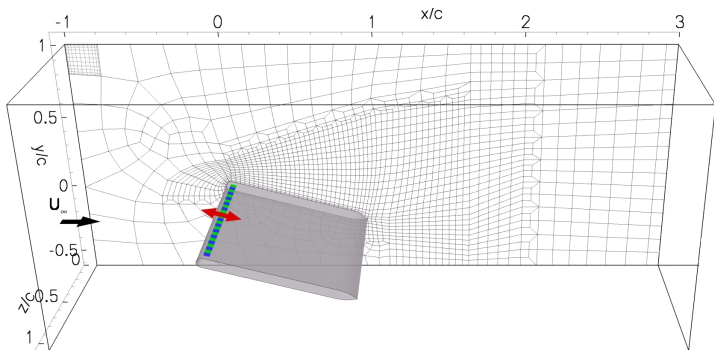
[T. Albrecht, J. Stiller, Proc. 3rd iTi Conf. Turbulence, 2008]

coupling advanced methods of physical numerical models and
methods of mathematical control

➔ improvement of flow qualities

aim: control transient/turbulent flows with high Reynolds numbers

- complex physical background → fine mesh



[T. Albrecht, J. Stiller, Proc. 3rd iTi Conf. Turbulence, 2008]

- appropriate small time steps
- for long-time simulation: high number of steps
- turbulence require a 3-D simulation

Semtex

- numerical simulation of fluid dynamics
- solve time dependent Navier-Stokes equations

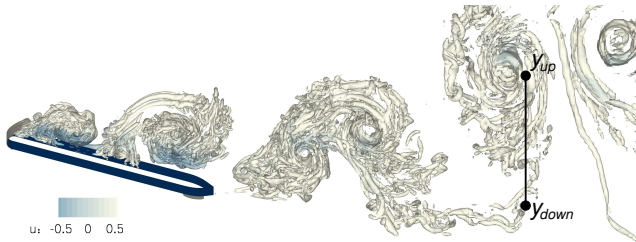
$$\frac{\delta u}{\delta t} + \nabla p = \nu \Delta u - (u \cdot \nabla)u + q$$

- spectral element method
 - high order finite element technique
 - combination of geometric flexibility of finite elements and high accuracy of spectral methods
- using:
 - parametrically mapped quadrilateral elements
 - Gauss-Lobatto-Legendre nodal shape function basis
 - continuous Galerkin projection

Modifications in Semtex

definition of a functional to minimize aerodynamic resistance

➔ minimize the momentum thickness using force q



$$J(x, q) = \sum_{n=1}^N \int_{t_{n-1}}^{t_n} \int_{y_{down}}^{y_{up}} \frac{u}{u_0} \left(1 - \frac{u}{u_0}\right) dy dt + \mu \|q\|^2 \rightarrow \min!$$

$x = (u, v)$ u_0 inflow velocity in horizontal direction

Flow control

optimization by using body forces

➔ high dimensional optimal control problem

➔ calculate gradients by using adjoint informations
to reduce the effort

Adjoint calculation

different techniques: algorithmic differentiation

discretization of continuous adjoint

hand coded

...

Gradient Calculation (Time Integration)

- 1 Forward integration (flow simulation)

$$x_{i+1} = F_i(x_i, q_i), \quad i = 1, \dots, l$$

with $x_i \in \mathbb{R}^n$ state, q_i control at time t_i

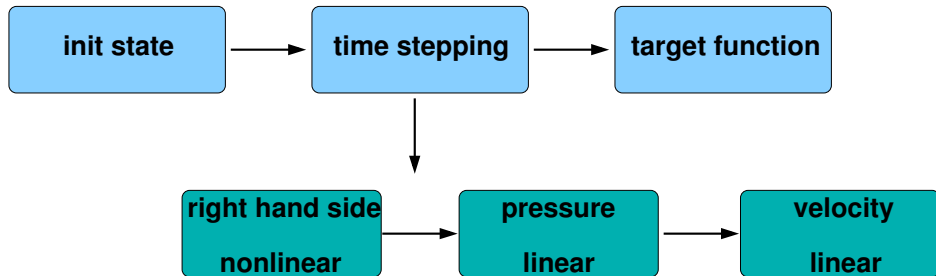
number of time steps l known

- 2 Evaluation of **target function** $J(x, q)$

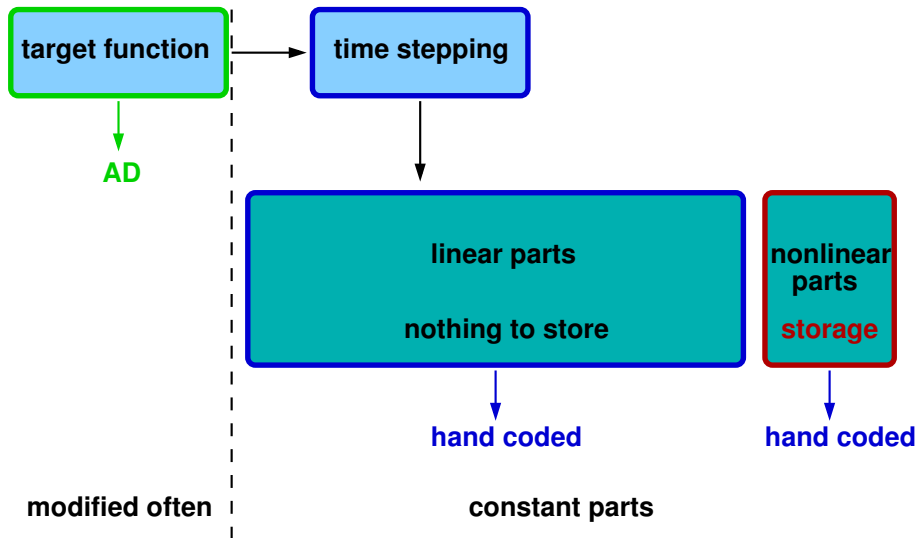
- 3 **Reverse integration:**

$$\bar{x}_{i-1} = \bar{F}_i(\bar{x}_i, \mathbf{x}_{i-1}, q_{i-1}, q_i), \quad i = l, \dots, 1$$

- 4 Gradient calculation



Adjoint?



Algorithmic Differentiation (AD)

Differentiation of “computer programs” within machine precision

Basic forms: [Griewank/Walther 2008]

forward mode: $\text{OPS}(F'(x)\dot{x}) \leq c_1 \text{OPS}(F), c_1 \in [2, 5/2]$

reverse mode: $\text{OPS}(\bar{y}^\top F'(x)) \leq c_2 \text{OPS}(F), c_2 \in [3, 4]$

$\text{MEM}(\bar{y}^\top F'(x)) \sim \text{OPS}(F)$

combination: $\text{OPS}(\bar{y}^\top F''(x)\dot{x}) \leq c_3 \text{OPS}(F), c_3 \in [7, 10]$

Tasks:

Derivatives (of any order)

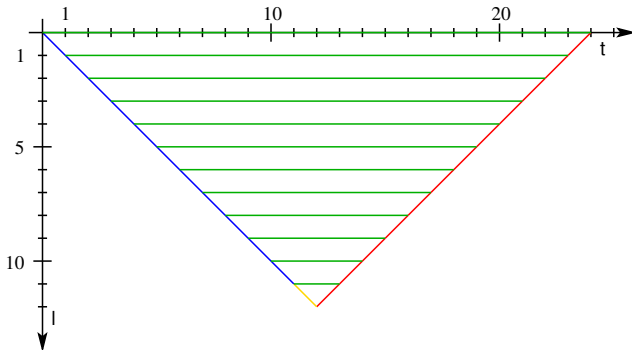
Important topics:

Advanced AD?

(sparsity patterns, structure exploitation, time-stepping, ...)

Store-Everything Approach

Example: 12 time steps



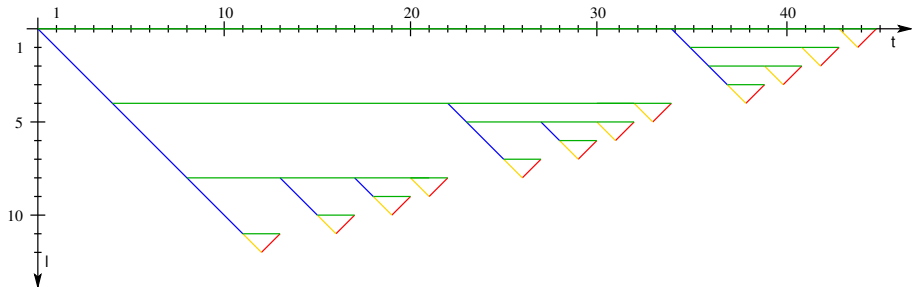
→ MEM = $O(l)$ $t = 2l$ **here:** #eval(F_i) = 12 $t = 24$

Binomial Checkpointing

Example: 12 time steps, 3 checkpoints and reusage of all checkpoints!

Binomial Checkpointing

Example: 12 time steps, 3 checkpoints and reuse of all checkpoints!



➔ MEM = c $t = ?$ **here:** $c=3$ #eval(F_i) = 21 $t=45$

Theorem (Complexity Result (Binomial Checkpointing))

Given

l = *total number of time steps to be reversed*

c = *number of checkpoints that can be used*

r = *unique integer satisfying*

$$\beta(c, r - 1) < l \leq \beta(c, r) \equiv \binom{c+r}{c}$$

➔ *Minimal number of forward steps required is given by*

$$t(c, l) = rl - \beta(c + 1, r - 1) + 1$$

Proof: Based on induction (Griewank, Walther 2000)

Numerical results

simulation of flows with $Re = 10^4$:

- spectral element mesh with 1757 elements
- polynomial order $N = 12$

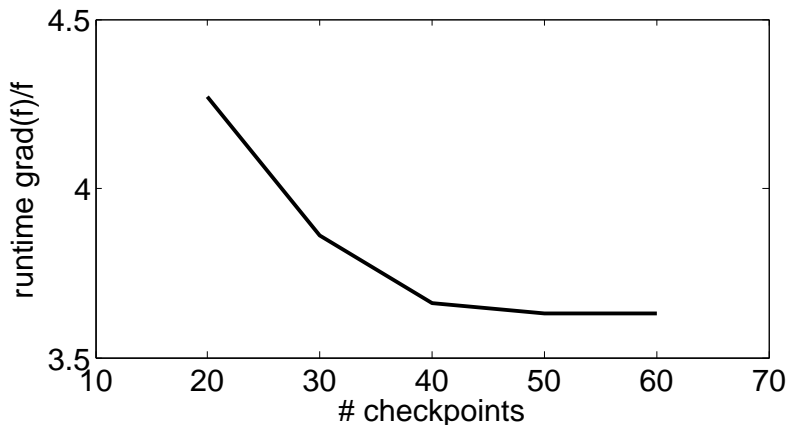
➔ acceptable level of divergence errors

execution time of one time step: 6.13s

execution time of one adjoint time step: 7.92s

➔ ratio: 1.3

Comparison of flow simulation runtime with adjoint calculation with 500 steps for 20, 30, 40, 50 and 60 checkpoints:



Outlook

Goal: adjoint-based control of turbulent flows
using electromagnetic fields

Applications: aerodynamics, crystal growth

Done:

- extension of a target function for optimization
- inclusion of hand coded adjoints and AD applications
- coupling with optimal checkpointing

Then:

- optimal control and experiments
- development of optimization strategies for turbulent flows